

NAG 5-1-75
IN-89-CR
317943
p.8

Image Construction from the IRAS Survey and Data Fusion*

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August 21, 1990

Abstract

The IRAS survey data can be used successfully to produce images of extended objects. The major difficulties, viz. non-uniform sampling, different response functions for each detector, and varying signal-to-noise levels for each detector for each scan, have been resolved. The results of three different image construction techniques are compared: co-addition, constrained least-squares, and maximum entropy. The maximum entropy result is superior. We present an image of the galaxy M51 with an average spatial resolution of 45 arc seconds, using 60 micron survey data. This exceeds the telescope diffraction limit of 1 minute of arc, at this wavelength. Data fusion is a proposed method for combining data from different instruments, with different spatial resolutions, at different wavelengths. Direct estimates of the physical parameters, temperature, density and composition, can be made from the data without prior image (re-)construction. An increase in the accuracy of these parameters is expected as the result of this more systematic approach.

1 Introduction

The Infrared Astronomical Satellite (IRAS) surveyed about 95% of the sky, in four broad spectral bands centred on 12, 25, 60, and 100 microns, during a ten month period in 1983. Precession at a rate of about 1° per day, kept the orbit of the spacecraft remaining perpendicular to the earth-sun vector (Figure 1). A semi-overlapping scan strategy was used for the 'all-sky' survey. Redundant coverage on the time scale of hours was provided by advancing the instrument in elongation by half of the width of the focal plane on a subsequent scan, usually the next orbit. For each spectral band there were two detector arrays (Figure 2). The arrays were arranged such that the second one scanned the same area of sky some 5 to 10 seconds later than the first. Different scans over the same area usually intersect at an angles due to precession of the satellite orbit. In addition, most of the scans were taken along small circles. Therefore, even small areas of the sky can be very unevenly covered.

*Talk presented at the Workshop on Restoration of HST Images and Spectra, August 21-22 1990, Baltimore.

The nature of the IRAS data is a collection of detector scans cross-cutting the sky at various angles. Consequently, the data are not in the form of an image and require non-traditional methods of image reconstruction. The term ‘image reconstruction’ actually does not apply; the procedure should be called *image construction*, because the true scene is unknown. Traditional reconstruction routines start with images on an evenly spaced rectilinear grid, of objects convolved with a single point-spread function.

Each of the 59 active IRAS detectors have different Response Functions (RFs). Most detectors have rectangular apertures of 0.75–3 (in the in-scan direction) to 4.5–5 arc minutes (in the cross-scan direction). Consequently, the spatial resolution is different in the two directions. It is possible to improve on the spatial resolution in the cross-scan direction because of the confirmation strategy, and because the two rows of detectors for each wavelength band are shifted by half a detector length in the cross-scan direction.

The detector outputs of two consecutive scans over the galaxy M51 at 60 microns wavelength are shown in Figure 3. This figure can be regarded as ruled-surface plots of the area. The signal is shifted between the two plots due to the half focal plane offset between scans. The main galaxy and its satellite NGC 5195 are clearly resolved. Figure 4 shows the centres of the individual sample positions of the M51 area, together with the outline of a standard 60 microns detector.

2 Data representation

Suppose datum d_n is the calibrated value of the n -th sample, taken by detector number i with its centre at the position (x_n, y_n) . Detector i has response function $R_i(x, y)$. Note that in general the RF profile is rotated according to the scan angle, as in Figure 4. Assume that for each sample this rotation has been taken into account in a temporary re-definition of $R_i(x, y)$, now with axes parallel to the axes of the desired map. The measured datum is now the result of a two dimensional integral of the sky brightness $b(x, y)$ and $R_i(x, y)$:

$$d_n = \int dx dy R_i(x_n - x, y_n - y) b(x, y) + n_n, \quad (1)$$

with n_n the noise in this n -th datum. Strictly, this is not a convolution since d_n is a single number, and not a function in the continuous variables x and y ; therefore it is called a sample.

Digitizing the brightness distribution in pixels reduces the integral to a summation:

$$d_n = \sum_{m=1}^M r_{nm} b_m + n_n, \quad (2)$$

with

$$r_{nm} = \int_{\text{area pixel } m} dx dy R_i(x_n - x, y_n - y). \quad (3)$$

Equation 2 can be interpreted as one equation in M unknown discrete brightnesses b_m , with known coefficients r_{nm} . There are in total N samples falling (partly) inside the map, forming a set of N equations in M unknowns (e.g. in Figure 4 $N \simeq 700$). This can be written as the matrix equation

$$d = R_{NM} b + n, \quad (4)$$

with $d = (d_1, \dots, d_N)$ and $n = (n_1, \dots, n_N)$, the data and the noise vector, respectively. The matrix R_{NM} is the $N \times M$ response matrix, in which the n th row is the set of coefficients R_{nm} for datum d_n . The unknown sky is represented by $b = (b_1, \dots, b_M)$, a vector of length M , obtained through stacking of the rows (or columns) of the desired map. The RFs are normalized to unit volume.

For samples which overlap the map boundary, it is assumed that the mean intensity just outside the map is the same as just inside the boundary. In this way boundary effects can be apodized.

Formally, the standard deviation n_n of each datum d_n is separately represented in Equation 4. The noise level is estimated from the available data, viz. signal plus noise, using a zero-sum filter. Differences in noise level of a factor three for the same detector, in consecutive orbits have been measured.

Summarizing, the non-uniform sampling, the different RF for each detector and their arbitrary rotations are incorporated in the response matrix. The different noise levels per detector and per scan are represented in the noise vector. The solution b is defined on a regular pixel grid. Since the individual response functions are used, no special provisions are necessary for the small detectors. The image construction has become a numerical mathematical problem which can be solved by many different methods.

Three methods have been compared, co-addition, constrained least-squares, and maximum entropy (see Bontekoe et al., 1991). The best results are obtained with maximum entropy, and are presented here.

3 Image construction

Even when the response matrix R_{NM} is perfectly known, the recovery of the original scene is mathematically impossible. Given the data d , the solution b is not unique. The problem is called *ill-posed*. In addition, the image (re-)construction problem is also *ill-conditioned*. The ‘signal-to-noise’ ratio of the final map is usually orders of magnitude worse than of the input data. Nevertheless, astronomically meaningful images can be obtained through regularization of the problem; maximum entropy is such a regularization.

In addition to solving for \hat{b} the standard deviation $\sigma_{\hat{b}}$ is also computed for each pixel, and serves as an error map. Although this ignores the covariance of neighbouring pixels, we use this error map to give us some indication of the reliability of features in the solution \hat{b} .

The images are constructed on a grid of 60×60 pixels, for the same map area as Figure 4 (15×15 arc minutes). Since there are $N \simeq 700$ samples and $M = 3600$ unknowns, the system of equations is underdetermined.

The MEMSYS3 software package of Skilling (1989) and Gull (1989) is used to obtain a maximum entropy estimate \hat{b} from M51 data. Until recently, the common procedure was to maximize the entropy under the constraint that the goodness-of-fit statistic equals the number of data points. This approach, however, has two major drawbacks. First, within this frame-work there is no consistent method to estimate the error map $\sigma_{\hat{b}}$, although the maximum entropy solution \hat{b} can be found straightforwardly. Second, the $\chi^2 = N$ criterion gives no allowance for the fact that significant data will generate structure in the image \hat{b} . This structure is like a set of parameters being fitted from the data. The effective number of parameters G supporting the underlying the structure should be

subtracted from N to obtain the number of degrees of freedom with which to compare χ^2 (see also the discussion in Gull, 1989).

The Bayesian estimate for the present case is $G \simeq 400$. The remaining $N - G \simeq 300$ is the number of degrees of freedom for this problem, and this should be used for the χ^2 statistic.

The maximum entropy solution shows fine details, especially in areas of low brightness where the other methods fail. Although such maps can be used sensibly, inferences ought to be made by overlaying them with masks and computing integrals over the masked solution.

The Figures 5 and 6 are the result of coarsening the maximum entropy solution, originally computed on 60×60 pixels, to an effective 30×30 grid by using 900 2×2 pixels boxcar masks, being 1 in the square of interest and 0 outside. The 60×60 grid is retained, however, and each group of four pixels is assigned the same value \hat{b} . Finally, this coarsened map is smoothed again using a 2×2 boxcar filter.

Division of brightness map by the error map yields Figure 6.

Most of the spiral structure in Figure 5 coincides well with the spiral arms of the $H\alpha$ image at 8 arc second resolution in van der Hulst et al. (1988). Extensions of the spiral arms at low brightness levels in the maximum entropy solution line up very well with outer parts of the spiral arms in $H\alpha$. The brightest point does not coincide with the nucleus in the $H\alpha$, but lies about 25 arc seconds to the South. The second source in the centre also has no visual counterpart and coincides with an inter-arm region. The companion galaxy seems resolved into a strong point source, towards its South-East, which coincides exactly with a sharp maximum in $H\alpha$ and extended structure towards the North and West.

Figure 6 summarizes the maximum entropy result in a statistical sense. The peaks in the galaxy and companion are 8σ and 9σ detections, respectively. The spiral arms are $1-3\sigma$ detections, and the secondary peak in the nucleus of M51 a 3σ detection. A unidentified source near the Northern boundary of the map is a 4σ detection, although its brightness is low. This might be an artifact from the treatment of the boundary.

A difficult issue is the final spatial resolution in the map. Since MEMSYS3 finds $G \simeq 400$, it is tempting to distribute this number evenly over the image as the number of independent picture elements, yielding an average spatial resolution of 45 seconds of arc. This is an improvement above the diffraction limit of the telescope as well as the classical limit imposed by sampling theory, both of which are 1 arc minute. Spatial resolution, however, is dependent on the 'signal-to-noise' in the original data and consequently non-uniform over the map. Therefore the 400 fitted parameters can not be evenly distributed, and areas of high brightness can have a better spatial resolution than average. The reverse is true for low brightness areas.

Although the IRAS survey mission was not intended to produce images, the major difficulties, viz. non-uniform sampling, different RFs for each detector, and different signal-to-noise levels for each detector for each scan, have been dealt with. Astronomically meaningful images can be produced of the far-infrared sky from the IRAS survey data, but they have to be constructed by advanced numerical techniques. Overall, the MEMSYS3 result is superior to the others. The images produced by MEMSYS3 show better spatial resolution, are non-negative and show plausible structure even at low brightness levels.

4 Data Fusion

One of the central questions in astrophysics is the correct identification of radiative processes operating in a given source. A fresh attack is proposed on the classification of physical processes in a complex field from a fundamentally new direction. The goal is to take data of a given field, obtained at a variety of wavelengths and spatial resolutions, and produce *images* of operative *physical* processes and the corresponding parameters.

First the data acquisition process is briefly described. Emphasis is given to the correct dimensions of all quantities. The sky brightness $B(\theta, \phi, \lambda)$ (in $Wm^{-2}m^{-1}sr^{-1}$) is a function of position of the sky (θ, ϕ) and of wavelength λ . When the instrument is pointed in the direction (θ_i, ϕ_i) , the response function, representing the blur, is defined $R(\theta, \phi, \theta_i, \phi_i)$ (dimensionless). In addition the blurred signal must pass a colour filter $F(\lambda)$ (dimensionless). Each datum of measured flux d_i (in Wm^{-2}) is now the result of both ‘convolutions’,

$$d_i = \int \int d\theta d\phi R(\theta, \phi, \theta_i, \phi_i) \int d\lambda F(\lambda) B(\theta, \phi, \lambda). \quad (5)$$

The important fact to notice is that the data acquisition is assumed to be linear with the input signal, which is the essential task for the calibration. It is this linearity that allows decomposition of the sky brightness into various components. Note also the difference between B here and b of the previous sections; the latter is the brightness integrated over the colour filter. The effect of noise is well understood in linear problems, and does not affect the theoretical analysis.

The ensemble of number densities $n_s(r, \theta, \phi, T_s)$ (in m^{-3}) of sources, indexed with s , can be estimated, when the source’s radii R_s (in m), and template spectra $I_s(\lambda, T_s)$ (in $Wm^{-2}m^{-1}sr^{-1}$) are given. The best fitting temperature T_s (in degrees K) of the various n_s results as part of the solution. An infinitesimal volume $dv = r^2 dr d\Omega$ (in m^3), radiates with a spectral power per unit wavelength dw (in Wm^{-1}) equal to

$$dw(r, \theta, \phi) = dv \sum_s n_s(r, \theta, \phi, T_s) 4\pi R_s^2 I_s(\lambda, T_s), \quad (6)$$

assuming isotropic radiation by all sources. The contribution to the sky brightness of dv is

$$dB(\theta, \phi, \lambda) = \frac{dw(r, \theta, \phi)}{4\pi r^2 d\Omega} e^{-\tau(r, \lambda)}, \quad (7)$$

where $\tau(r, \lambda)$ is the optical depth (dimensionless). Integrating over the line of sight, the sky brightness becomes

$$B(\theta, \phi, \lambda) = \int_0^\infty dr e^{-\tau(r, \theta, \phi, \lambda)} \sum_s n_s(r, \theta, \phi, T_s) R_s^2 I_s(\lambda, T_s). \quad (8)$$

Under the condition that the absorption, represented by τ , and the emission I_s are independent of the local radiation field, the sky brightness B is a linear function of the densities n_s .

Since the data acquisition is assumed linear with brightness, there is linear relation between the data d_i and n_s , the physical parameters of the distribution of matter in space. It requires simultaneous solution of the equation of radiative transfer and the instrumental inversion. However complicated, the problem is linear, and linear problems can be solved by many numerical techniques.

The critical factor in resolving three-dimensional structures in the universe is knowledge of the factor $\tau(r, \theta, \phi, \lambda)$. The full solution $n_s(r, \theta, \phi, T_s)$ for all sources describes the composition, temperature, and density of all visible matter. If this information is not present in the data, upper bounds for the desired densities are still a useful result.

If one is less confident in the quality of the data or one's knowledge of the absorption, one can apply a two dimensional version of the theory in which only projected densities are defined.

Acknowledgement

The author acknowledges support from NASA grant NAG 5 1246.

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Figure captions

Figure 1: Schematic drawing of IRAS orbital geometry.

Figure 2: IRAS focal plane. Of the 62 infrared detectors the 3 filled-in were inoperative; the cross-hatched detectors showed a higher noise level. The normal scan direction of images is shown.

Figure 3: Detector scans from two consecutive scans covering the galaxy M51 (60 microns). The scan direction is from right to left and each has a length of 0.5 degree. The detector scans are displaced vertically corresponding to their cross-scan position in the focal plane. The figures can be regarded as ruled surface plots. The maximum signal in both plots is 430 detector units.

Figure 4: Positions of the samples in the area of M51 (60 microns). The circles represent the two smallest detectors in the band, viz. detectors 11 and 31 (see Fig. 1). The scans roughly run from top-left to bottom-right. The area is covered by about 700 samples. The fat contours outline the RF profile of a normal size detector, at the 90, 50, 10, 2, and 1% level of its maximum. The detector is centred at the position indicated with an asterisk. A grid with a one arc minute spacing is superimposed.

Figure 5: Map of M51 (60 microns) from the MEMSYS3 maximum entropy method. Area is the same as in Figure 4. The lowest contour is at a level 150; subsequent contours with increments of a factor 2.

Figure 6: Signal-to-noise map of the MEMSYS3 solution of M51. Lowest contour is at 1σ , subsequent contours are separated by 1σ .

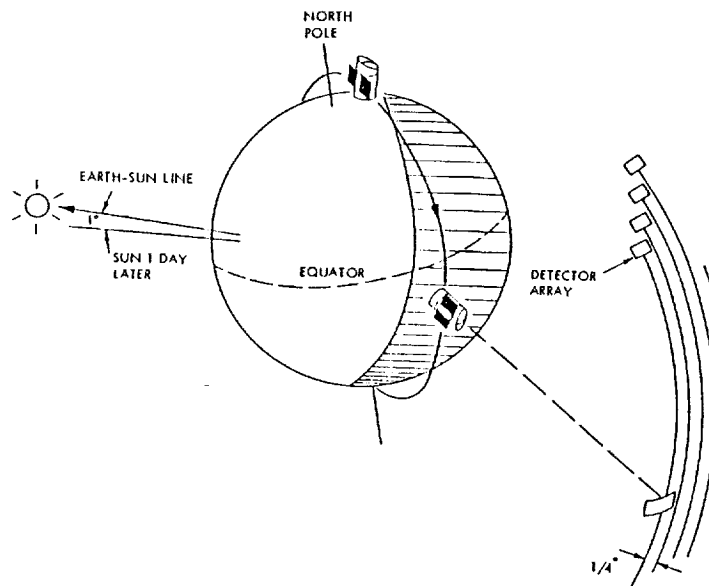


Fig. 1

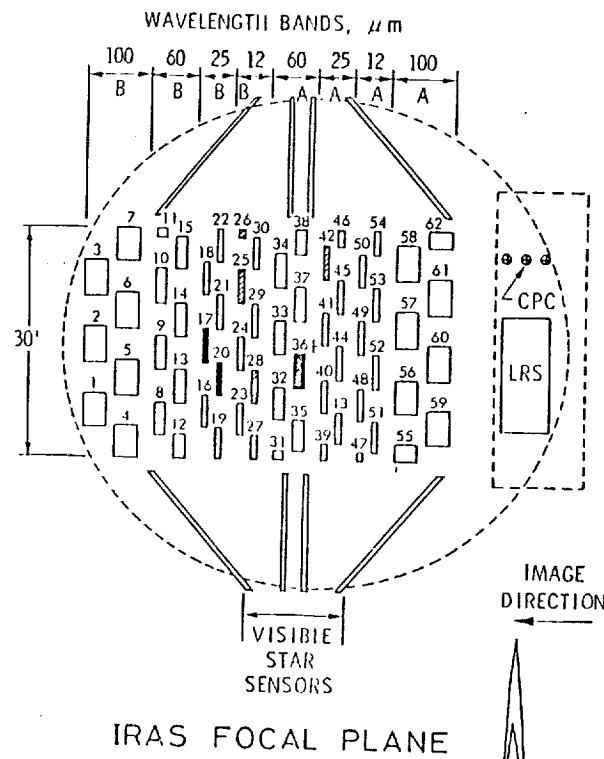


Fig. 2

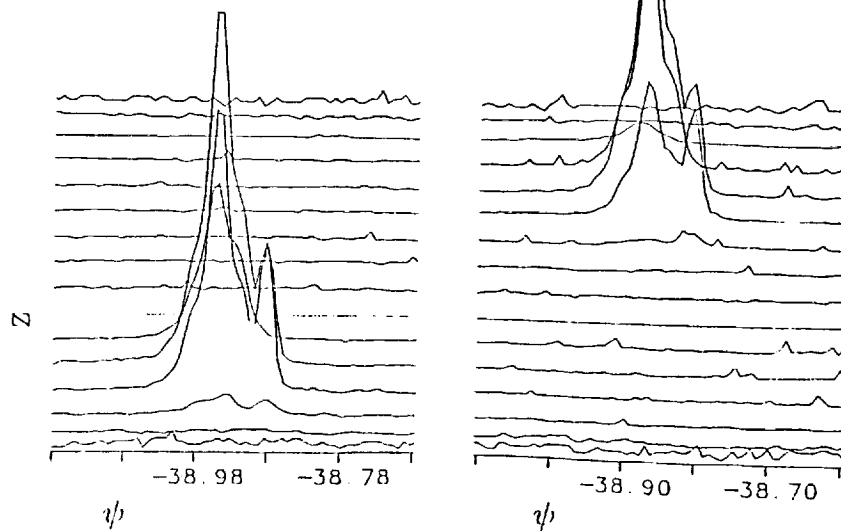


Fig. 3

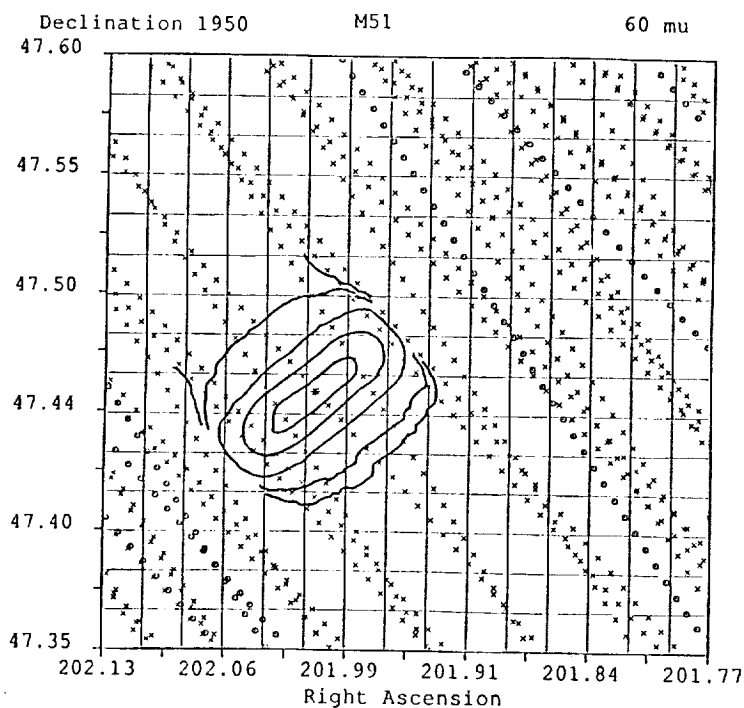


Fig. 4.

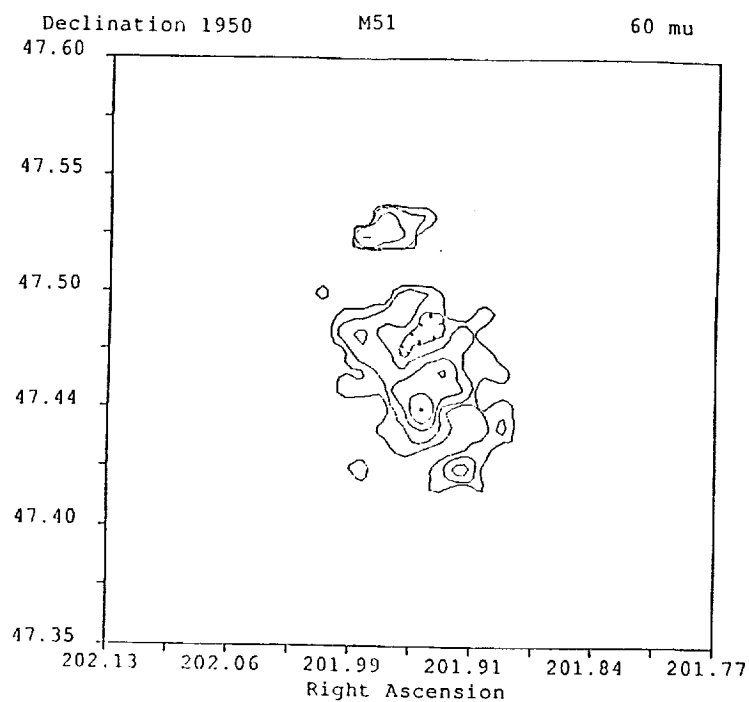


Fig. 5

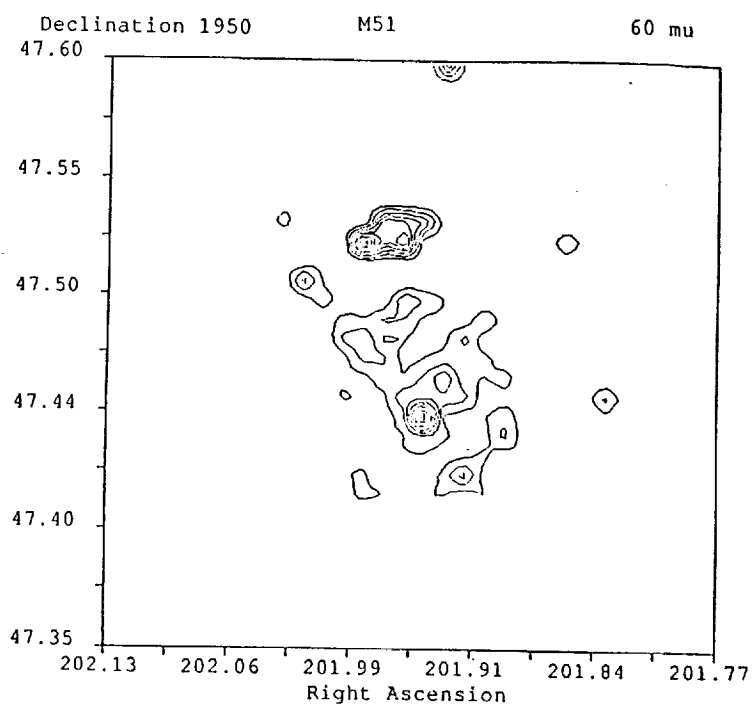


Fig. 6